

# Distributed Processing and Energy Saving Techniques in Mobile Crowd Sensing

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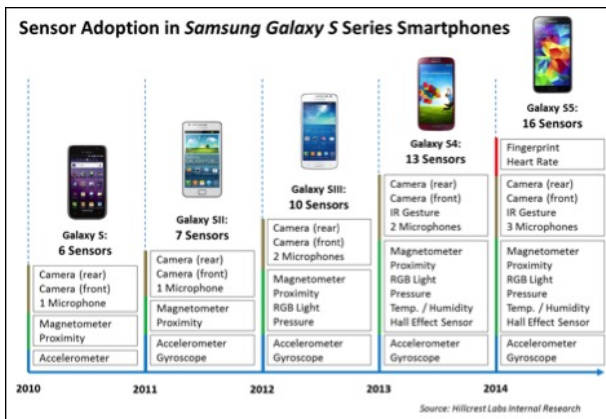
# Overview

- 1 Motivation
- 2 Research Gap
- 3 Our Proposal
- 4 Implementation
- 5 Conclusions

# Mobile Crowd Sensing

- Successful society and city management relies on efficient monitoring of urban and community dynamics for decision and policy making
- However, commercial sensor network techniques have never been successfully deployed in the real world due to several reasons, such as
  - High installation cost
  - Insufficient spatial coverage
- MCS is a large-scale sensing paradigm based on the power of user-companioned devices, including mobile phones, smart vehicles, wearable devices, and so on [1]

# Mobile Crowd Sensing

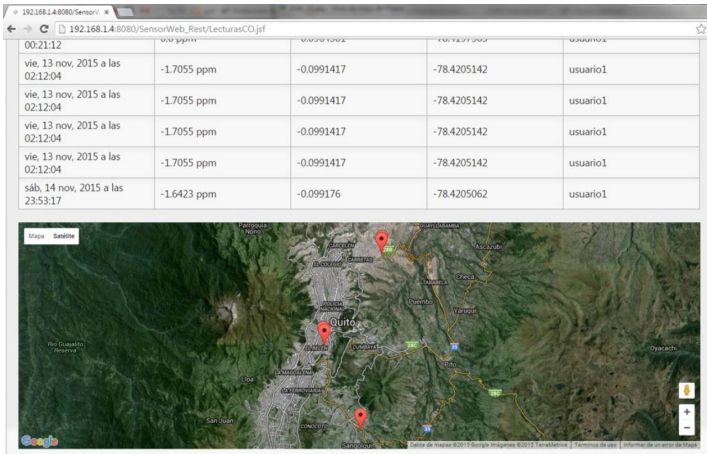


Typical smartphone sensors

# Projects

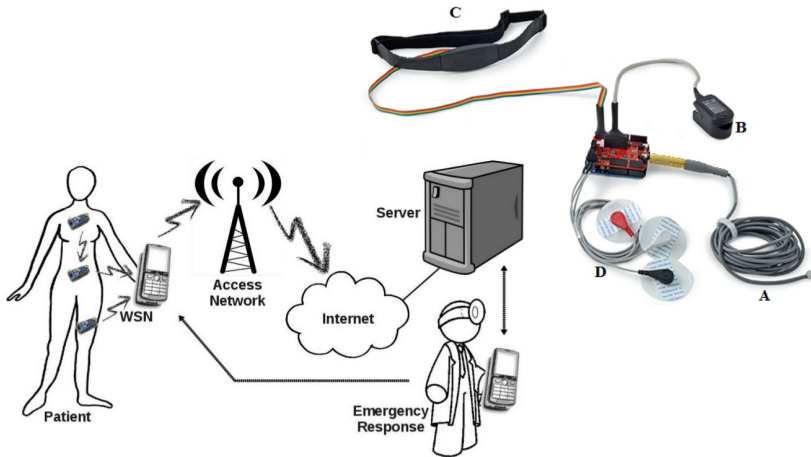
- The information must be aggregated in the cloud for large-scale sensing and community intelligence mining
- Previous projects:
  - Real-time environmental monitoring using smartphone apps
  - Healthcare monitoring of elderly people [2]
- Current projects:
  - Public safety monitoring through an Integrated Security Service
  - Earthquake monitoring based on smartphone sensors [3]

# Projects



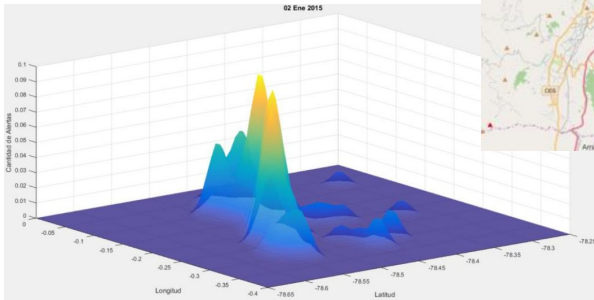
## Environment monitoring [4]

# Projects



Healthcare monitoring [5]

# Projects



Public safety monitoring



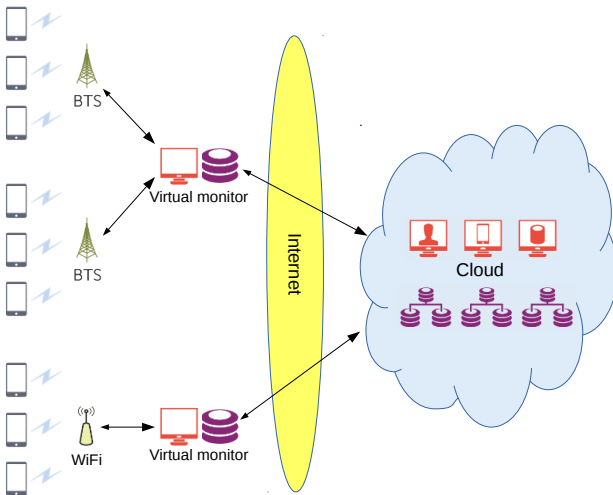
# Research Gap

- Deploying MCS applications in large-scale environments is not a trivial task [6]
  - Heterogeneity of sensing hardware and mobile platforms
  - Increasing network bandwidth demands of emerging crowd sensing apps (*i.e.*, high data rate sensors)
  - Real-time processing is challenging because of high latencies
  - Participating users are exposed to a significant drain on limited mobile battery resources
- Proposed strategies produce [7]:
  - High user engagement
  - Poor data quality
  - Excessive energy consumption

# Distributed Processing

- The dominant approach of aggregating all data to a single data center inflates the timeline of analytics [8]
- In fact, MCS normally relies on an Internet-scale searchable repository (*i.e.*, centralized analytics)
- Our proposal includes a *hierarchical distributed architecture* in order to:
  - Generate very low and predictable latencies
  - Solve scalability issues
- Then, the challenge is running queries over geo-distributed inputs by optimizing placement of both data and tasks of queries [9]

# Distributed Processing



Hierarchical distributed architecture

# Energy Saving

- In other words, we are looking for a *self-adaptive edge analytics* by pushing processing to the edge [10]
- However, mobile devices operates on a finite supply of energy contained in its battery
- In order to reduce energy costs without sacrificing precision of data, our strategy is combining:
  - Piggybacking based on smartphone app opportunities
  - Lightweight compression of data
  - Simple anomaly and outlier detection [11]

# Methodology

- Distributed processing simulation:
  - Virtual smartphones sending fake accelerometer measurements
  - Virtual monitors receive sensor data and apply some analytics
  - Centralized server keeps summary of recorded activity  
*(HPC cluster: Linux, 320 GB, 730 cores – up to 4096 nodes)*
- Smartphone energy consumption characterization [12]:
  - Hardware (NI myDAQ)
  - Software (Android apps)

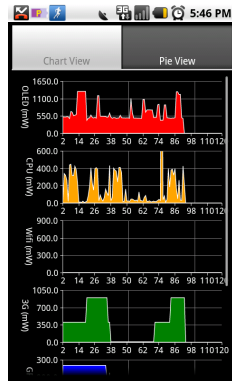
# Evaluation

- Distributed processing simulation:
  - Throughput supported by a single virtual monitor
  - Experienced latency among tiers
  - Performance of distributed queries
- Smartphone energy consumption:
  - Energy consumption of activated sensors
  - Energy consumption in function of sensor data rate
  - Energy consumption of data compression algorithms
  - Energy consumption of simple anomaly detection algorithms

# Evaluation



NI myDAQ



PowerTutor

# Conclusions

- We have motivated MCS — a cross-space, heterogeneous crowdsourced sensing paradigm for large-scale sensing and computing
- MCS will foment and enhance numerous application areas, such as environment monitoring, intelligent transportation, urban sensing, mobile social recommendation, and so on
- However, the deployment of MCS applications in large-scale environments is a challenging task
- Then, we are proposing a hierarchical distributed architecture where processing is pushed to the edge without increasing energy consumption of battery-operated devices



# Conclusions

- For evaluating our proposal, we are implementing a simulation of the distributed architecture plus actual power consumption measurements in smartphones
- We expect to get our first results once the simulated platform is working at the end of September 2016
- After that, we are going to start working in enhancements to our original proposal
- There are many research opportunities related to distributed processing and energy consumption issues in the field of MCS

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